

Math 151b: Problem Set 7

Owen Jones

3/6/2024

Problem 1

(a) Case 1: $\mathbf{v} = \mathbf{0}$

Trivially $\|A\mathbf{v}\| = \|\mathbf{0}\| = 0 \leq 0 = \|A\|0 = \|A\| \cdot \|\mathbf{v}\|$

Case 2: $\mathbf{v} \neq \mathbf{0}$

Let $\mathbf{w} = \frac{\mathbf{v}}{\|\mathbf{v}\|} \Rightarrow \|\mathbf{w}\| = 1$.

$\|A\mathbf{w}\| \leq \max_{\|\mathbf{x}\|=1} \|A\mathbf{x}\| = \|A\| \Rightarrow \|A\mathbf{w}\| \leq \|A\|$.

Thus, $\|A\mathbf{w}\| = \|A \frac{\mathbf{v}}{\|\mathbf{v}\|}\| = \frac{1}{\|\mathbf{v}\|} \|A\mathbf{v}\| \leq \|A\| \Rightarrow \|A\mathbf{v}\| \leq \|A\| \cdot \|\mathbf{v}\|$.

(b) Suppose $\|\mathbf{x}\| = 1$.

Observe $\text{Range}(\mathbf{B}) \subset \mathbb{R}^n$.

$\|AB\mathbf{x}\| \leq \|A\| \cdot \|B\mathbf{x}\| \leq \|A\| \cdot \|B\| \cdot \|\mathbf{x}\| = \|A\| \cdot \|B\|$ by part (a).

Since this inequality holds for any $\|\mathbf{x}\| = 1$, the inequality must hold for $\|AB\| = \max_{\|\mathbf{x}\|=1} \|AB\mathbf{x}\| \leq \|A\| \cdot \|B\|$.

Thus, $\|AB\| \leq \|A\| \cdot \|B\|$.

Problem 2

We will prove $\rho(\cdot)$ is not a matrix norm by contrapositive. Let $\mathbf{A} \in \mathbb{R}^{n \times n} \neq \mathbf{0}$ be a strictly triangular matrix. Because the eigenvalues of a triangular matrix are the elements along the main diagonal, $\lambda_i = 0, i = 1, \dots, n$. It follows $\rho(\mathbf{A}) = \max_{1 \leq i \leq n} |\lambda_i| = 0$. Because $\rho(\mathbf{A}) = 0 \not\Rightarrow \mathbf{A} = \mathbf{0}$, $\rho(\cdot)$ is not a matrix norm.

Problem 3

$$G_j = D^{-1}(L + U), G_g = (D - L)^{-1}U$$

$$D = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}, L = \begin{bmatrix} 0 & 0 & 0 \\ -1 & 0 & 0 \\ 1 & 1 & 0 \end{bmatrix}, U = \begin{bmatrix} 0 & 1 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix}$$

$$G_j = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 1 & -1 \\ -1 & 0 & -1 \\ 1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ -1 & 0 & -1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 0 & 0 \\ 1 & 1 & 0 \\ -1 & -1 & 2 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & 1 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$G_g = \begin{bmatrix} \frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & 1 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 1 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ 0 & -\frac{1}{2} & \frac{1}{2} \\ 0 & 0 & -\frac{1}{2} \end{bmatrix}$$

(a) For G_j $\det(\lambda I - G_j) = \lambda^3 + \frac{5}{4}\lambda$, $\lambda = 0, \frac{\sqrt{5}i}{2}, -\frac{\sqrt{5}i}{2}$
 $\Rightarrow \rho(G_j) = |\frac{\sqrt{5}i}{2}| = \frac{\sqrt{5}}{2} < 1$

(b) For G_g eigenvalues are elements along diagonal, $\lambda = 0, -\frac{1}{2}$
 $\Rightarrow \rho(G_g) = |-\frac{1}{2}| = \frac{1}{2} < 1$

Problem 4

(a) $G_j = D^{-1}(L + U) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{bmatrix}$
For G_j $\det(\lambda I - G_j) = \lambda(\lambda^2 - 2) - 2(\lambda - 2) + 4(\lambda + 1) = \lambda^3$
 $\Rightarrow \rho(G_j) = |0| = 0 < 1$

(b) $(D - L)^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & 2 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} =$
 $\begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix}$
 $G_g = (D - L)^{-1}U = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 0 & -2 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & -2 & 2 \\ 0 & 2 & -3 \\ 0 & 0 & 2 \end{bmatrix}$
For G_g eigenvalues are elements along diagonal, $\lambda = 0, 2$
 $\Rightarrow \rho(G_g) = |2| = 2 > 1$

Problem 5

(a) Proof by induction:
Base case: $\|\mathbf{x}^{(1)} - \mathbf{x}\| = \|(G\mathbf{x}^{(0)} + c) - (G\mathbf{x} + c)\|$
 $= \|G(\mathbf{x}^{(0)} - \mathbf{x})\| \leq \|G\| \|\mathbf{x}^{(0)} - \mathbf{x}\|$ by Problem 1.
Induction hypothesis: Assume for some k $\|\mathbf{x}^{(k)} - \mathbf{x}\| \leq \|G\|^k \|\mathbf{x}^{(0)} - \mathbf{x}\|$.
Induction step: $\|\mathbf{x}^{(k+1)} - \mathbf{x}\| = \|(G\mathbf{x}^{(k)} + c) - (G\mathbf{x} + c)\|$
 $= \|G(\mathbf{x}^{(k)} - \mathbf{x})\| \leq \|G\| \|\mathbf{x}^{(k)} - \mathbf{x}\|$
 $\leq \|G\|^{k+1} \|\mathbf{x}^{(0)} - \mathbf{x}\|$ by the induction hypothesis.
Hence, by induction, the claim holds for all k .

(b) (i) Proof by induction:
Base case: $\|\mathbf{x}^{(2)} - \mathbf{x}^{(1)}\| = \|(G\mathbf{x}^{(1)} + c) - (G\mathbf{x}^{(0)} + c)\|$

$$= \|G(\mathbf{x}^{(1)} - \mathbf{x}^{(0)})\| \leq \|G\| \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \text{ by Problem 1.}$$

Induction hypothesis: Assume for some k $\|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\| \leq \|G\|^{k-1} \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|$.

$$\begin{aligned} \text{Induction step: } \|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\| &= \|(G\mathbf{x}^{(k)} + c) - (G\mathbf{x}^{(k-1)} + c)\| \\ &= \|G(\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)})\| \leq \|G\| \|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\| \\ &\leq \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \text{ by the induction hypothesis.} \end{aligned}$$

Hence, by induction, the claim holds for all k .

(ii) Proof by induction:

Base case: the base case is proved in part (a). We know $\|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\| \leq \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|$

Induction hypothesis: Assume for some $m > k$

$$\|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\| \leq \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k-1} \|G\|^i$$

$$\begin{aligned} \text{Induction step: } \|\mathbf{x}^{(m+1)} - \mathbf{x}^{(m)}\| &\leq \|G\|^m \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \text{ by part (a).} \\ \|\mathbf{x}^{(m+1)} - \mathbf{x}^{(m)}\| + \|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\| &\leq \|G\|^{m-k} \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| + \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k-1} \|G\|^i \\ &= \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k} \|G\|^i. \end{aligned}$$

By the triangle inequality $\|\mathbf{x}^{(m+1)} - \mathbf{x}^{(k)}\| \leq \|\mathbf{x}^{(m+1)} - \mathbf{x}^{(m)}\| + \|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\|$

$$\text{Thus, } \|\mathbf{x}^{(m+1)} - \mathbf{x}^{(k)}\| \leq \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k} \|G\|^i.$$

Hence, by induction, the claim holds for all m .

(iii) Suppose $\|G\| < 1$.

$$\|\mathbf{x}^{(m)} - \mathbf{x}\| \leq \|G\|^m \|\mathbf{x}^{(0)} - \mathbf{x}\| \text{ from part (a).}$$

It follows $\lim_{m \rightarrow \infty} \|\mathbf{x}^{(m)} - \mathbf{x}\| = 0 \Rightarrow \mathbf{x}^{(m)} \rightarrow \mathbf{x}$ as $m \rightarrow \infty$.

$$\text{Thus, } \|\mathbf{x}^{(k)} - \mathbf{x}\| = \lim_{m \rightarrow \infty} \|\mathbf{x}^{(k)} - \mathbf{x}^{(m)}\|.$$

$$\|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\| \leq \|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k-1} \|G\|^i \text{ from part (b)(ii).}$$

$$\|G\|^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\| \sum_{i=0}^{m-k-1} \|G\|^i = \frac{\|G\|^k - \|G\|^m}{1 - \|G\|} \text{ for all } m > k \text{ by the sum of a geometric series.}$$

$$\text{Using } \|G\| < 1, \lim_{m \rightarrow \infty} \frac{\|G\|^k - \|G\|^m}{1 - \|G\|} = \frac{\|G\|^k}{1 - \|G\|}.$$

Hence, by the squeeze theorem, $\|\mathbf{x}^{(k)} - \mathbf{x}^{(m)}\| \leq \frac{\|G\|^k - \|G\|^m}{1 - \|G\|}$ for all

$$m > k \text{ implies } \|\mathbf{x}^{(k)} - \mathbf{x}\| \leq \frac{\|G\|^k}{1 - \|G\|}.$$

(Note: I know the squeeze theorem requires the inequality to hold for every element of the sequence, but we can easily get around that by defining 2 sequences $A_n := \|\mathbf{x}^{(k)} - \mathbf{x}^{(k+1+n)}\|$ and $B_n := \frac{\|G\|^k - \|G\|^{k+1+n}}{1 - \|G\|}$ which satisfy the condition $A_n \leq B_n$ for all $n \geq 0$).

Problem 6

$$\begin{aligned} \mathbf{x}_j^{(1)} &= G_j \mathbf{x}_j^{(0)} + c_j = c_j = \begin{bmatrix} \frac{3}{5} \\ \frac{2}{5} \\ -\frac{11}{8} \\ -\frac{11}{5} \end{bmatrix} \\ \mathbf{x}_j^{(2)} &= G_j \mathbf{x}_j^{(1)} + c_j = (G_j + I)c_j = \begin{bmatrix} 1 & -\frac{1}{2} & & \\ -\frac{1}{2} & 1 & & \\ & \frac{1}{2} & \frac{2}{5} & \\ & & \frac{1}{5} & 1 \end{bmatrix} \begin{bmatrix} \frac{3}{5} \\ \frac{2}{5} \\ -\frac{11}{8} \\ -\frac{11}{5} \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} \\ \frac{3}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \end{bmatrix} \end{aligned}$$

$$\mathbf{x}_g^{(1)} = G_g \mathbf{x}_g^{(0)} + c_g = c_g = \begin{bmatrix} \frac{3}{5} \\ \frac{11}{5} \\ -\frac{11}{40} \\ -\frac{451}{200} \end{bmatrix}$$

$$\mathbf{x}_g^{(2)} = G_g \mathbf{x}_g^{(1)} + c_g = (G_g + I) c_g = \begin{bmatrix} 1 & -\frac{1}{2} & & \\ 0 & \frac{4}{5} & \frac{2}{5} & \\ 0 & \frac{1}{8} & \frac{1}{25} & \frac{1}{40} \\ 0 & \frac{1}{40} & \frac{1}{25} & \frac{41}{40} \end{bmatrix} \begin{bmatrix} \frac{3}{5} \\ \frac{11}{5} \\ -\frac{11}{40} \\ -\frac{451}{200} \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} \\ \frac{66}{25} \\ -\frac{539}{1600} \\ -\frac{18139}{8000} \end{bmatrix}$$

Answers!!

$$\mathbf{x}_j^{(1)} = \left[\frac{3}{5} \quad \frac{5}{2} \quad -\frac{11}{8} \quad -\frac{11}{5} \right]^\top$$

$$\mathbf{x}_j^{(2)} = \left[-\frac{13}{20} \quad \frac{33}{20} \quad -\frac{2}{5} \quad -\frac{99}{40} \right]^\top$$

$$\mathbf{x}_g^{(1)} = \left[\frac{3}{5} \quad \frac{11}{5} \quad -\frac{11}{40} \quad -\frac{451}{200} \right]^\top$$

$$\mathbf{x}_g^{(2)} = \left[-\frac{1}{2} \quad \frac{66}{25} \quad -\frac{539}{1600} \quad -\frac{18139}{8000} \right]^\top$$

(b) See Python code

(c) The Gauss-Seidel method took half as many iterations to converge within the prescribed tolerance of 10^{-6} .

Jacobian method: 4115 iterations

GS method: 1983 iterations

```
In [1]: import numpy as np
        from numpy import linalg
        from matplotlib import pyplot as plt
        from math import sqrt
```

```
In [2]: def LDU(A):
        D=np.diag(np.diagonal(A))
        L=-1*(np.tril(A)-D)
        U=-1*(np.triu(A)-D)
        return D,L,U
```

```
In [3]: def Jacobi(A,b,x_0,tol=1e-6,n_max=10000):
        D,L,U=LDU(A)
        n=0
        D_inv=linalg.inv(D)
        G_j=np.matmul(D_inv,L+U)
        c_j=D_inv.dot(b)
        while linalg.norm(A.dot(x_0)-b,np.inf)>=tol and n<n_max:
            x_0=G_j.dot(x_0)+c_j
            n+=1
        return x_0,linalg.norm(A.dot(x_0)-b,np.inf),n
```

```
In [4]: def GS(A,b,x_0,tol=1e-6,n_max=10000):
        D,L,U=LDU(A)
        n=0
        D_L_inv=linalg.inv(D-L)
        G_g=np.matmul(D_L_inv,U)
        c_g=D_L_inv.dot(b)
        while linalg.norm(A.dot(x_0)-b,np.inf)>=tol and n<n_max:
            x_0=G_g.dot(x_0)+c_g
            n+=1
        return x_0,linalg.norm(A.dot(x_0)-b,np.inf),n
```

```
In [5]: def BVP(f,a,b,c,alpha,beta,numpts):
        xvec=np.linspace(a,b,numpts+1)
        h=xvec[1]-xvec[0]
        bvec=f(xvec[1:-1])
        bvec[0]=bvec[0]-alpha/h**2
        bvec[-1]=bvec[-1]-beta/h**2
        A=-1*(2/h**2+c)*np.identity(numpts-1)+np.diag((1/h**2)*np.ones(numpts-2),k=1)
        return A,bvec
```

```
In [6]: f= lambda x:x*x0
```

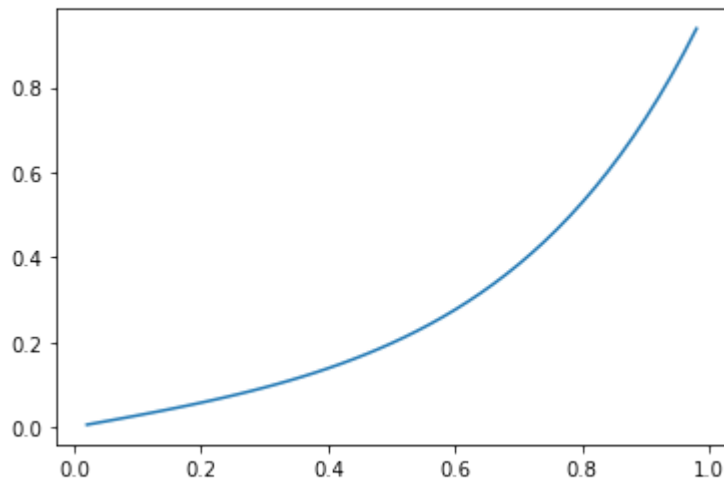
```
In [14]: A,b=BVP(f,0,1,10,0,1,50)
```

```
In [15]: u_j,res_j,n_j=Jacobi(A,b,np.zeros(np.shape(A)[0]))
```

```
In [16]: u_g, res_g, n_g=GS(A,b,np.zeros(np.shape(A)[0]))
```

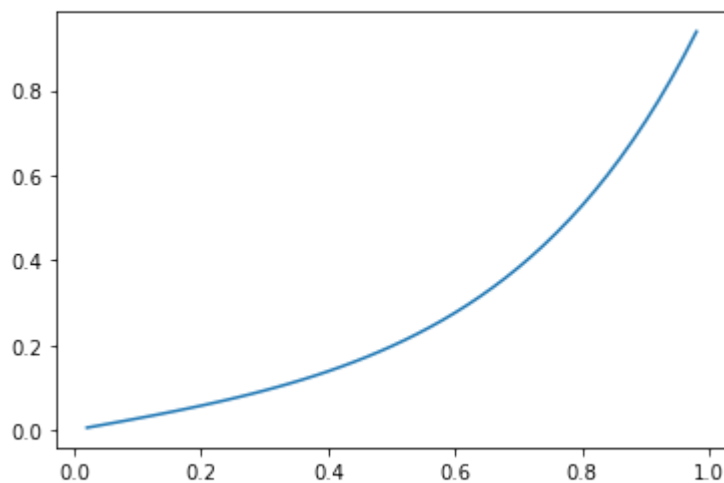
```
In [17]: plt.plot(np.linspace(0,1,len(u_j)+2)[1:-1],u_j)
```

```
Out[17]: [<matplotlib.lines.Line2D at 0x7fdba83d2c40>]
```



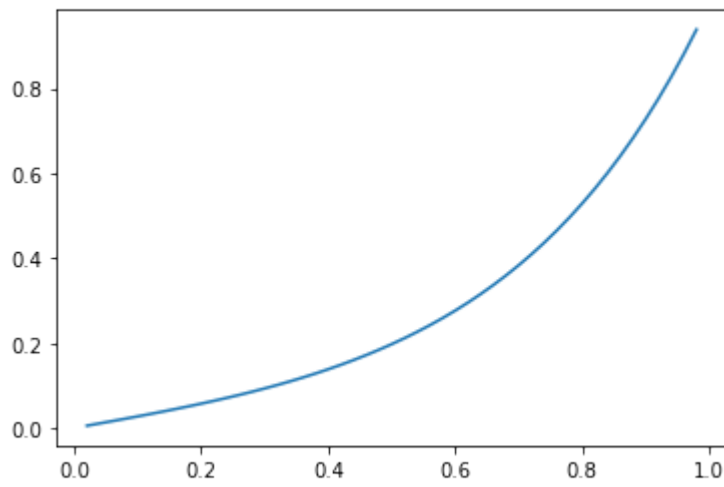
```
In [18]: plt.plot(np.linspace(0,1,len(u_g)+2)[1:-1],u_g)
```

```
Out[18]: [<matplotlib.lines.Line2D at 0x7fdc487adeb0>]
```



```
In [19]: plt.plot(np.linspace(0,1,np.shape(A)[0]+2)[1:-1],linalg.inv(A).dot(b))
```

```
Out[19]: [<matplotlib.lines.Line2D at 0x7fdc4888e310>]
```



In [20]: `n_j`

Out[20]: 4115

In [21]: `n_g`

Out[21]: 1983

In []: